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Multimomics resolution of molecular events during a day in the life of *Chlamydomonas*

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The unicellular green alga *Chlamydomonas reinhardtii* displays metabolic flexibility in response to a changing environment. We analyzed expression patterns of its three genomes in cells grown under light–dark cycles. Nearly 85% of transcribed genes show differential expression, with different sets of transcripts being up-regulated over the course of the day to coordinate cellular growth before undergoing cell division. Parallel measurements of select metabolites and pigments, physiological parameters, and a subset of proteins allow us to infer metabolic events and to evaluate the impact of the transcriptome on the proteome. Among the findings are the observations that *Chlamydomonas* exhibits lower respiratory activity at night compared with the day; multiple fermentation pathways, some oxygen-sensitive, are expressed at night in aerated cultures; we propose that the ferredoxin, FDX9, is potentially the electron donor to hydrogenases. The light stress-responsive genes *PSBS*, *LHCSR1*, and *LHCSR3* show an acute response to lights-on at dawn under abrupt dark-to-light transitions, while *LHCSR3* genes also exhibit a later, second burst in expression in the middle of the day dependent on light intensity. Each response to light (acute and sustained) can be selectively activated under specific conditions. Our expression dataset, complemented with coexpression networks and metabolite profiling, should constitute an excellent resource for the algal and plant communities.

photobioreactor | systems biology | cell division | histone expression | chloroplast

Life evolved under the constant pressure of limited resources. The invention of oxygenic photosynthesis by early cyanobacteria partially relieved these limitations, placing biology under the dependence of the Sun. This had the unexpected benefit of bringing predictability to physiology and metabolic pathways. The rotation of the Earth around the Sun allowed the temporal separation of incompatible biochemical reactions over the course of a day. On a global scale, much of transcription was placed early under the control of molecular oscillators or circadian clocks to properly gate key physiological events to the right time of day (or night) and optimize resource allocation.

Over the course of evolution, the rhythmic fraction of the transcriptome has adjusted to the physiology and habitat of the organism. Recent technical advances, such as microarrays, transcriptomics, and untargeted proteomics have brought unprecedented detail to our understanding of circadian and diurnal rhythms in daily physiology and behavior. However, the intrinsic biology of more complex eukaryotic model systems comes with drawbacks: they are composed of several tissues whose individual signals are diluted at the whole-organism level. The strength of rhythms can also be affected by developmental stage.

In this regard, unicellular algae are well suited for studies of rhythmic transcription in response to a changing environment (1). Among them, *Chlamydomonas reinhardtii* is a workhorse in the fields of photosynthesis, chloroplast biology, ciliopathies, and

metal homeostasis, and is a reference organism in biotechnology for production of high-value bio-products and biofuel (2, 3). *Chlamydomonas* is also metabolically flexible, with phototrophic, heterotrophic, and fermentation capabilities, the latter predominating under anoxic conditions that prevail in the dark (4). It can acclimate to changes in light intensity (5, 6). Located at the base of the green lineage, *Chlamydomonas* shares fundamental regulatory and metabolic pathways with other algae and land plants, and possesses the advantages of a unicellular microbial system that can be (i) grown in large volumes and (ii) easily synchronized with alternating light–dark cycles, both in terms of diurnal gene expression and cell division. Most of the algal primary metabolism, protein synthesis, DNA replication, and organelle biogenesis pathways are diurnally coordinated to sustain growth toward cell division (7, 8). *Chlamydomonas* cells can undergo several consecutive divisions at dusk, producing 2 to 32 daughter cells

Significance

Chlamydomonas reinhardtii is the premier reference organism for understanding unicellular green algae. *Chlamydomonas* is an important model for photosynthesis as well as fermentation and other anaerobic pathways under dark anoxic conditions. We have produced a diurnal transcriptome, validated by sub-proteomic analyses, and matched with measurements of pigments, select metabolites, and physiological parameters. We report that the majority of the algal genome is differentially expressed over the course of the day and the timing of specific genes is dictated by their biological function. We also discovered that fermentation rather than respiration is the preferred metabolic fate of starch-derived glycolytic pyruvate. We offer our rich dataset to the algal and plant communities.

Author contributions: D.S. and S.S.M. designed research; D.S., S.S., S.D.G., P.A.S., S.O.P., C.D.N., T.M.-A., E.S., and G.J.B. performed research; S.D.G., S.O.P., C.D.N., T.M.-A., E.S., A.P.M.W., M.S.L., and G.J.B. contributed new reagents/analytic tools; D.S., S.S., S.D.G., P.A.S., S.O.P., C.D.N., T.M.-A., E.S., A.P.M.W., M.S.L., G.J.B., and S.S.M. analyzed data; and D.S., P.A.S., and S.S.M. wrote the paper.

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The authors declare no conflict of interest.

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Data deposition: Transcriptome data were deposited in the NCBI Gene Expression Omnibus (GEO) database, <https://www.ncbi.nlm.nih.gov/geo> (accession no. GSE112394). The mass spectrometry proteomics data have been deposited in the ProteomeXchange Consortium (www.proteomexchange.org) via the PRIDE partner repository (accession no. PXD010794).

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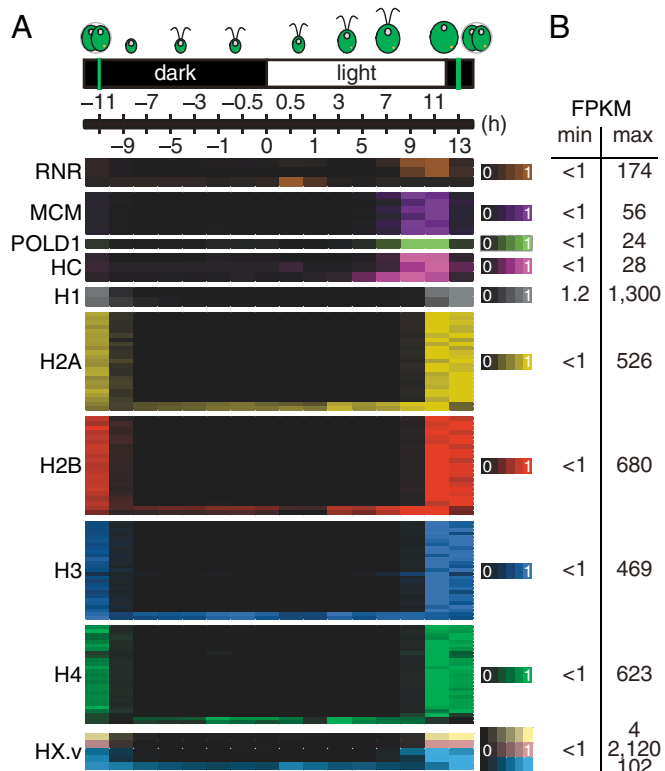


Fig. 2. High synchrony across cells illustrated by the expression of genes involved in DNA replication. (A) Heatmap showing transcript abundance for genes involved in DNA RNR (3 genes, brown), DNA helicase (MCM, 6 genes, purple), DNA polymerase (POLD1), histone chaperones (HC, three genes, magenta), histone H1 (2 genes, gray), histone H2A (22 genes, yellow), histone H2B (20 genes, red), histone H3 (26 genes, blue), and histone H4 (26 genes, green). Histone variants (HX.v) are shown in a lighter version of the color used to depict their respective histone groups. In all cases, expression was normalized to 1 for the highest value. (B) Dynamic expression range of replication genes. Minimum and maximum RNA abundance (in FPKM) from the average of three independent experiments is shown. The green vertical lines indicate the timing of cell division.

and animal systems. In *Chlamydomonas*, the algal orthologs for *CAF-1*, *ASF1*, and *HIRA* (*HIR1*) are coexpressed with peak expression between CT+9 and CT+11, a pattern more consistent with replication-dependent histone deposition. However, *ASF1* expression does rise transiently after dawn, and *HIR1* expression starts to increase around CT+3, which could allow both proteins to participate in nucleosome deposition outside of DNA replication (Fig. 2).

Most, but not all genes involved in DNA replication share the same expression pattern. For example, the second *Chlamydomonas* *RNR2* gene (*RIR2b*) is only expressed shortly after dawn (CT+0.5), which is matched by a secondary peak in *RNR1* expression, suggesting a distinct role, possibly in DNA damage repair in response to photodamage. *RNR1* protein is still present around CT+1, although at low levels, and can therefore associate with the morning-specific *RIR2b* subunit for RNR function (*SI Appendix, Fig. S3B*). Similarly, two genes per canonical core histone gene family and a single histone H3 variant are expressed constitutively outside the DNA replication window, and may provide an emergency pool of histones for epigenetic control in response to changes in the environment (Fig. 24). Taken together, these results demonstrate the very high degree of synchronization of our cell population. These results also highlight the potential for gene discovery and for distinguishing functions of paralogs from our dataset, which we further explore in the next sections.

The Majority of *Chlamydomonas* Genes Exhibit a Rhythmic Expression

Pattern. To parse out our rich dataset, we performed a principle component analysis (PCA) to reduce complexity, shown in Fig. 34. A testament to the controlled experimental conditions is the grouping of time points CT-11 and CT+13, which are temporally identical over a diurnal cycle. In addition, we observed coclustering of samples collected later during the night, from CT-7 to CT-0.5, during which cells remain relatively more metabolically quiescent during the G_0 phase of the cell cycle. The first two components capture over 73% of the total variance in expression: the first component separates samples between day and night, while the second component sorts them according to timing (Fig. 34). A PCA of the *Arabidopsis* transcriptome displayed a very similar pattern when seedlings were grown under light-dark cycles (22), suggesting that our samples may cycle robustly, from synchronized cell division, diurnal or circadian rhythms, or both.

To increase the resolution of the analysis, we next turned to *k*-means clustering, yielding 11,377 differentially expressed genes, or 85% of all transcribed genes, that fall in 16 clusters for nucleus-encoded genes, 3 clusters for chloroplast genes, and a single cluster for mitochondrial genes. These numbers are comparable to those published in an earlier diurnal study where *Chlamydomonas* cells were grown at high CO₂ (8). Visualizing the entire dataset as a heatmap emphasizes the degree to which the *Chlamydomonas* transcriptome is driven at a genome-wide level by diurnal oscillations (Fig. 3B). Because the timing of physiological and molecular events during the day is critical for cell growth and survival, we determined the time of peak expression, or phase, using the algorithm JTK_CYCLE (23). Updating the heatmap of the transcriptome sorted by phase rather than cluster essentially yields comparable results (Fig. 3C), and underscores the impact of diurnal rhythms on algal biology.

Functional Gene Clustering Around the Diurnal Cycle. As illustrated with DNA replication genes (Fig. 2), temporal coexpression of genes across the diurnal cycle can point to similar function. We surveyed expression patterns and potential timing coincidence for a number of functional categories: cilia [from CiliaCut (24)] and other flagella-associated proteins (FAPs), nucleus-encoded ribosomal protein genes (RPGs), mitochondrial and chloroplast electron transfer chains (ETCs), and the carbon-concentrating mechanism (CCM). The expression of most FAPs peaks in the middle of the night between CT-5 and CT-6, following the completion of DNA replication (Fig. 4) and histone deposition on the newly synthesized DNA (Fig. 4A), as expected. Cells will have a fully functional pair of cilia by next dawn and will be ready for phototaxis and optimal photosynthesis. Nucleus-encoded cytosolic and plastid RPGs peak early during the day to sustain photosynthesis and incorporate newly fixed carbon into proteins critical for cell growth. Chlorophyll biosynthetic genes all peak during the first half of the day, and precede the observed rise in cellular chlorophyll content (*SI Appendix, Fig. S4*).

As the number of genes within a category increases, so does the spread of measured phases. For example, many genes involved in the CCM show peak expression during the day (0–9 h after lights-on), while others reach peak levels either during the night or shortly before dawn (Fig. 4C). Genes encoding proteins found in the cilia proteome (25) make up another example: although FAPs exhibit a clear phase preference for the middle of the night, the larger set of genes encoding ciliary proteins takes on a much broader phase distribution. Many support cilia regrowth following cell division, but a significant fraction shows peak expression during the day or shortly after dusk (Fig. 4C). This includes 1 of the 12 *Chlamydomonas* carbonic anhydrases, *CAH6*. CAHs catalyze the interconversion of CO₂ into bicarbonate (HCO₃[−]) and protons. The expression pattern of *CAH6* is unique among CAHs: *CAH6* is low during the day, when other CAHs (*CAH1* to *-5*) reach their peak, and is much less induced by low CO₂ than other CAHs, like the mitochondria-localized *CAH4* and *CAH5* (SI Appendix, Fig. S5). *CAH6* localizes to the cilia, where it interacts with FAP12 (26) and may contribute to

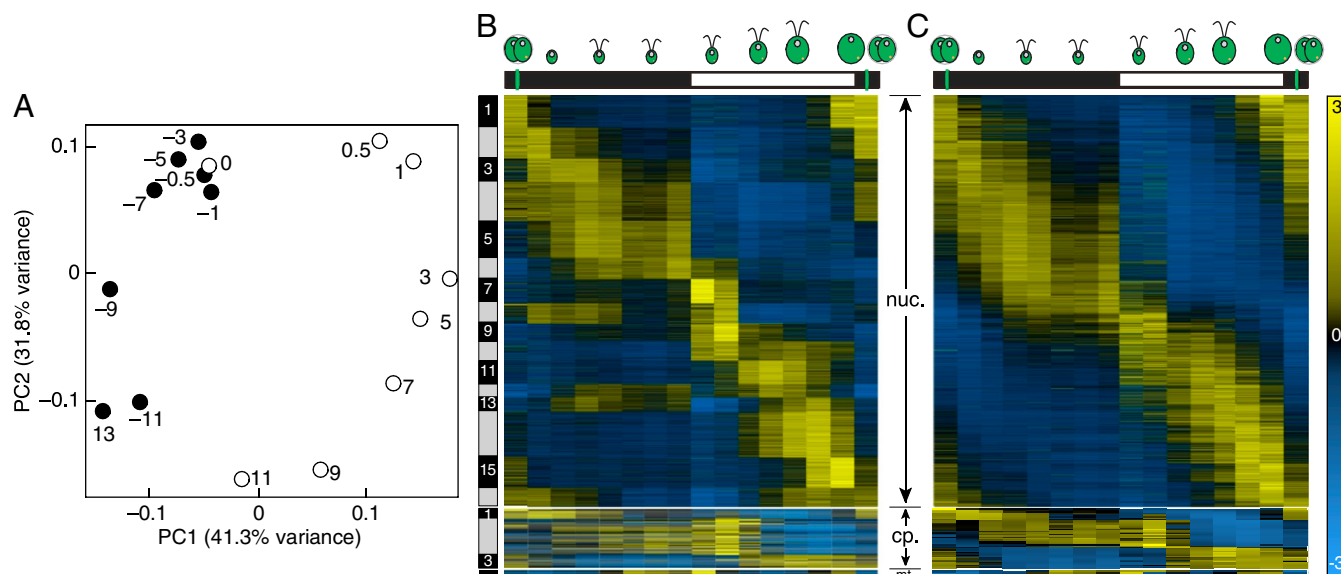


Fig. 3. The *Chlamydomonas* transcriptome is highly dynamic and rhythmic over one diurnal cycle. (A) PCA of the *Chlamydomonas* transcriptome over the diurnal cycle: samples collected at night (●), samples collected during the day (○). The time of sample collection is indicated next to the corresponding circle. (B) Heatmap representation of gene expression for 10,465 differentially expressed genes over the diurnal cycle. Genes were included if at least 1,000 counts were detected in the experiment, and there was a Benjamini–Hochberg-adjusted *P* value < 0.01 for differential expression. Genes were grouped into 16 clusters of nuclear genes, 3 clusters of chloroplast genes, and 1 mitochondrial cluster. For easier visualization, the vertical scale of the organellar transcriptomes (chloroplast transcriptome, *Middle*; mitochondrial transcriptome, *Bottom*) is enlarged 50 times relative to the nuclear transcriptome. (C) Heatmap representation of gene expression for 6,916 genes based on peak time (phase) of expression, as determined by the algorithm JTK_CYCLE, with minimal expression of 1 FPKM for at least one time-point. For easier visualization, the vertical scale of the organellar transcriptomes (chloroplast transcriptome, *Middle*; mitochondrial transcriptome, *Bottom*) is enlarged 25 times relative to the nuclear transcriptome. The green vertical lines indicate the timing of cell division.

chemotaxis toward HCO_3^- . Sensitivity of *Chlamydomonas* cells to HCO_3^- is under diurnal control with highest sensitivity at night, coincident with peak *CAH6* expression (27).

Cells produce ATP via photosynthesis and respiration to fuel growth, for which they rely on their organelles. Most genes involved in photosynthetic electron transfer are expressed during

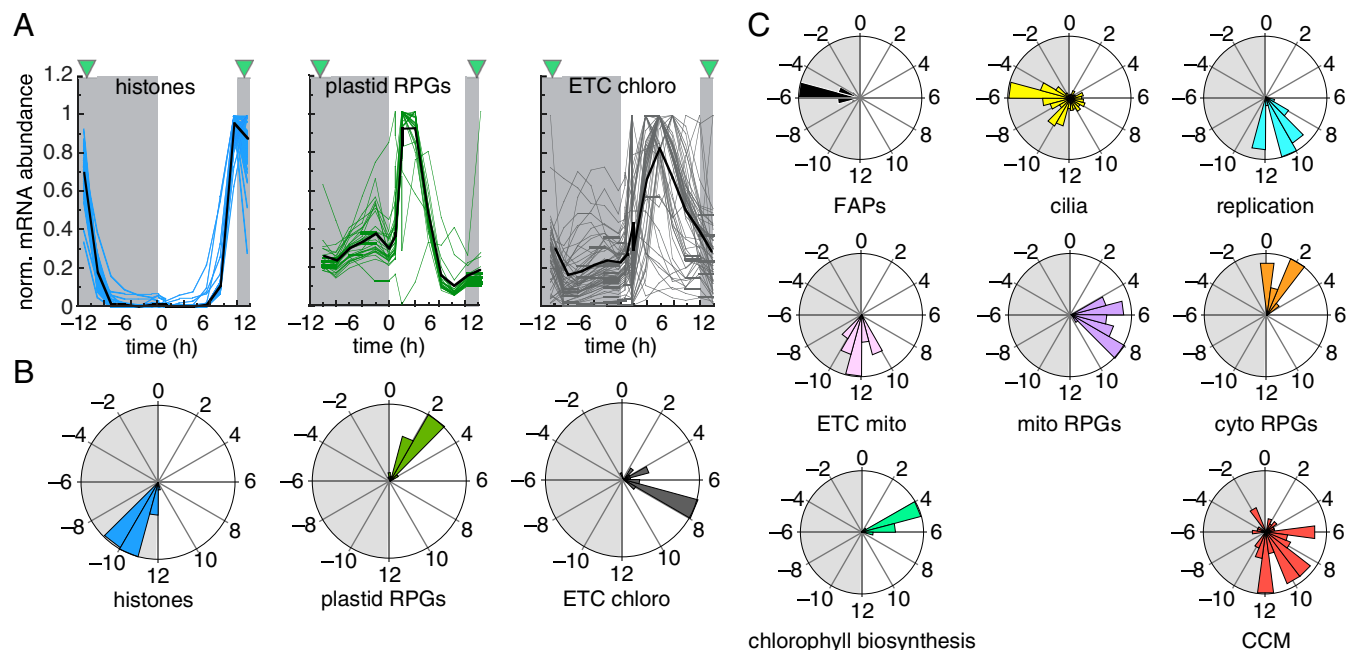


Fig. 4. Genes with similar function are coexpressed. (A) Examples of synchronous gene-expression profiles for histone genes, nucleus-encoded plastid RPGs, and genes involved in the chloroplast ETC. All expression estimates are normalized to 1 for the highest value. The green triangles indicate the timing of cell division. (B and C) Phase distributions over the diurnal cycle for genes belonging to chosen gene categories. Phase values are based on the algorithm JTK_CYCLE, with a cut-off rate BH.Q of 1×10^{-5} . Functional categories shown are: histones, DNA replication, FAPs, RPGs, mitochondrial and chloroplast ETC, chlorophyll biosynthesis, CCM, cilia (flagellar proteome, including all FAPs).

the day (Fig. 4C). As cells prepare to shut down photosynthetic processes in anticipation of darkness, mitochondria-driven respiration appears to take over, as the expression of its electron transport chain constituents reaches maximum expression around dusk (Fig. 4C).

Our results demonstrate clustered expression of genes that participate in similar cellular functions. Furthermore, cellular events proceed in an orderly fashion during the day and night parts of the diurnal cycle, as hinted by PCA (Fig. 3A). Finally, our dataset enables association of putative functions with distinct family members or unknown genes (SI Appendix, Fig. S6) and allows the generation of testable hypotheses. For example, the expression pattern of mitochondrial electron transport chain genes suggests that respiration should be higher at night, during which time cells break down starch produced during the day as a carbon source, which we set out to test next.

Respiration Does Not Reach Maximum Capacity at Night. We first determined when cells accumulate starch. Total organic carbon per cell varies about twofold over the diurnal cycle, reaching its lowest levels at the end of the night and rising gradually during the day (Fig. 5A). Starch increases fivefold over the course of the day, particularly in the later part of the day, and at its peak accounts for about 25% of total organic carbon in the cell at the end of the day (Fig. 5B). Starch levels decrease continuously during the night, and are almost depleted by the beginning of the next day, as is also the case in *Arabidopsis* (28, 29). We therefore assumed that respiratory activity would follow starch degradation; the cultures were continuously aerated, so oxygen should not be limiting. Indeed, the truncated hemoglobin *THB8*, a marker for strict hypoxia (30), is not expressed in our cultures (Dataset S10). However, oxygen consumption, a quantitative measure of respiration, is in fact low in the night, despite cellular inability to produce ATP from photosynthesis (Fig. 5C). The potential for respiration (or respiratory capacity) is high, but the components are not engaged (Fig. 5C). Going back to our expression dataset, transcripts for the genes encoding all complexes of the mitochondrial respiratory chain, including cytochrome *c* oxidase (complex IV) and ATP synthase (complex V), display two distinct peaks, one early at night, and one in the middle of the day (SI Appendix, Fig. S7).

We then analyzed the fraction of oxygen consumption directly attributable to cytochrome *c* oxidase or to alternative oxidases utilizing selective inhibitors, because oxygen consumption measurements reflect the combined activities of these enzymes. Each oxidase type accounts for half of total oxygen consumption during the day, while at night most of the oxygen consumption activity can

be attributed to cytochrome *c* oxidase (Fig. 5D). However, because cytochrome *c* oxidase operates at capacity during the day, but not in the night (Fig. 5C), cytochrome *c* oxidase-dependent O_2 consumption is still higher at each time point during the day compared with the night (Fig. 5D). Therefore, the increased oxygen consumption noted in the light period cannot be solely attributed to the alternative oxidases. We conclude that the high respiratory activity during the day reflects a greater cellular demand for ATP to facilitate maximal macromolecular metabolism and growth. During the dark portion of the diurnal cycle the demand for ATP is lower because the cells are in G_0 (the metabolically more quiescent stage of the cell cycle), and even low levels of respiration may be sufficient to fulfill cellular ATP requirements. Nevertheless, this left open the question of how glycolytically generated NADH might be reoxidized and the fate of pyruvate, which we explored next.

Fermentation as the Path for Reoxidation of the NADH Generated During Glycolysis. The oxidation of Glc to pyruvate generates NADH, which must be reoxidized to NAD^+ to support continued glycolysis. In aerobic organisms, this occurs via respiration, which is coupled to ATP synthesis. Although the yield of ATP from aerobic oxidation of NADH is substantial, the process is slower than anaerobic pathways for NADH reoxidation (e.g., conversion of pyruvate to lactate or ethanol). These anaerobic pathways are well-studied in *Chlamydomonas* (Fig. 6A). We obtained validation of the fermentation hypothesis in the expression levels of *Chlamydomonas* fermentation genes, and discovered some unexpected additional complexity. All genes reach their peak accumulation after dusk and are predominantly more highly expressed at night, with the exception of *LDH1* and *PFL1* (which show some expression also in the day period). In addition, individual genes appear to be turned off sequentially (Fig. 6B). For many fermentation genes, high expression is restricted to the night, suggesting potential light-mediated mRNA degradation or transcriptional repression. We hypothesize that the proteins encoded by these genes offer multiple options over the course of the dark part of the diurnal cycle for the reoxidation of NADH to NAD^+ . Genes encoding enzymes involved in the conversion of pyruvate to lactate, ethanol, and formate, including *PFL1*, show peak expression right after dusk followed by a fast decline, while genes encoding enzymes generating acetate (*PAT2* and *ACK1*) and CO_2 (*PDC3*) reach maximal expression later during the night and over a longer time scale (Fig. 6B). One of the products of pyruvate metabolism, lactate, is the sole soluble metabolite that accumulates at night (Fig. 6C), which further substantiates the operation of anaerobic routes for pyruvate metabolism.

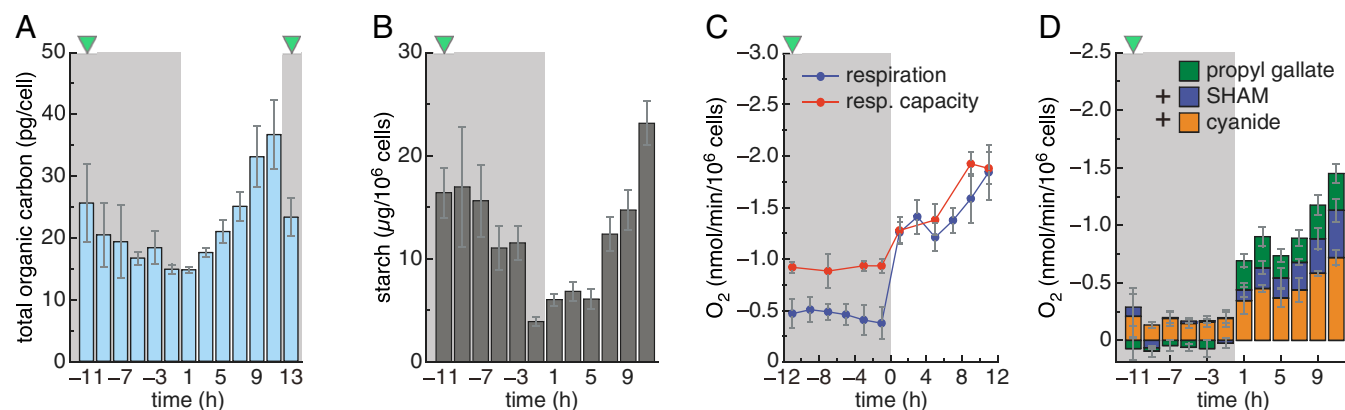


Fig. 5. *Chlamydomonas* cells do not use respiration to full capacity in the dark. Total nonpurgeable organic carbon (A) and starch content (B) of *Chlamydomonas* cells over the diurnal cycle. Data shown on a per cell basis. The green triangles indicate the timing of cell division. (C) Oxygen consumption (blue line) and respiration (resp) capacity of cells treated with the mitochondrial uncoupling agent FCCP (red line). Oxygen consumption was measured on the same samples before and after addition of FCCP. Data shown on a per cell basis. (D) Relative contribution of cytochrome *c* oxidase and alternative oxidases in oxygen consumption. Potassium cyanide (cyanide in figure) inhibits cytochrome *c* oxidase, while SHAM and propyl gallate target mitochondrial and plastid terminal oxidases (AOX and PTOX, respectively). Data shown on a per cell basis.

RNAs encoding enzymes that are not sensitive to oxygen begin to rise already by day's end, while RNAs encoding oxygen-sensitive enzymes like *PFR1* do not accumulate until after dusk (Fig. 6*B*). Genes encoding the other oxygen-sensitive enzymes, *PFL1* and *PAT2*, begin to be expressed during the day, but exhibit a large jump in their expression at dusk (421% for *PFL1* and 360% for *PAT2*, between CT+11 and CT+13). These gene-expression patterns are roughly mirrored in protein accumulation, assessed immunologically for *PFR1*, *PFL*, *ADH*, and *HYDA1/2* (Fig. 6*D*), and by proteomics analysis of soluble cell fractions for *PFR1*, *HYDG*, and *HYDA2* (SI Appendix, Fig. S8). Because RNA half-lives are typically much shorter than that of proteins, unless there is a specific mechanism to degrade or destabilize the proteins, the proteins persist into the day period. Some proteins are lost more rapidly, perhaps because loss of their oxygen-labile cofactors renders them susceptible to proteolytic digestion. Expression of the genes encoding pyruvate dehydrogenase enzymes *PDH1* and *PDH2* (catalyzing the oxidative degradation of pyruvate to acetyl-CoA) is

restricted to the middle of the day, preceding the accumulation of numerous TCA intermediates, as expected (Fig. 6*B* and *C*).

Our cultures were not anaerobic, but genes encoding anaerobic routes for pyruvate metabolism are expressed at or near the same levels in these experiments as they are in anaerobic cells (Fig. 6*B*), with the exception of *HYDA1* and *PFR1*, which are 10–20 times more highly expressed in strict hypoxia (31–34). We turned to other transcriptome datasets with samples collected in the dark to independently validate the hypothesis that anaerobic pyruvate catabolism represents the typical rather than exceptional metabolic program: we observed essentially the same pattern and abundance in an earlier diurnal time course, including the differential expression behavior of *HYDA1*, *PFR1*, and *ADH1* (8). Other studies also detected significant expression of fermentation genes in nonhypoxic cultures (35, 36); all indicate that expression of fermentation genes is not restricted to strict anaerobic conditions, and that aerated *Chlamydomonas* cultures have the potential to metabolize pyruvate by fermentation rather than via the TCA cycle at night.

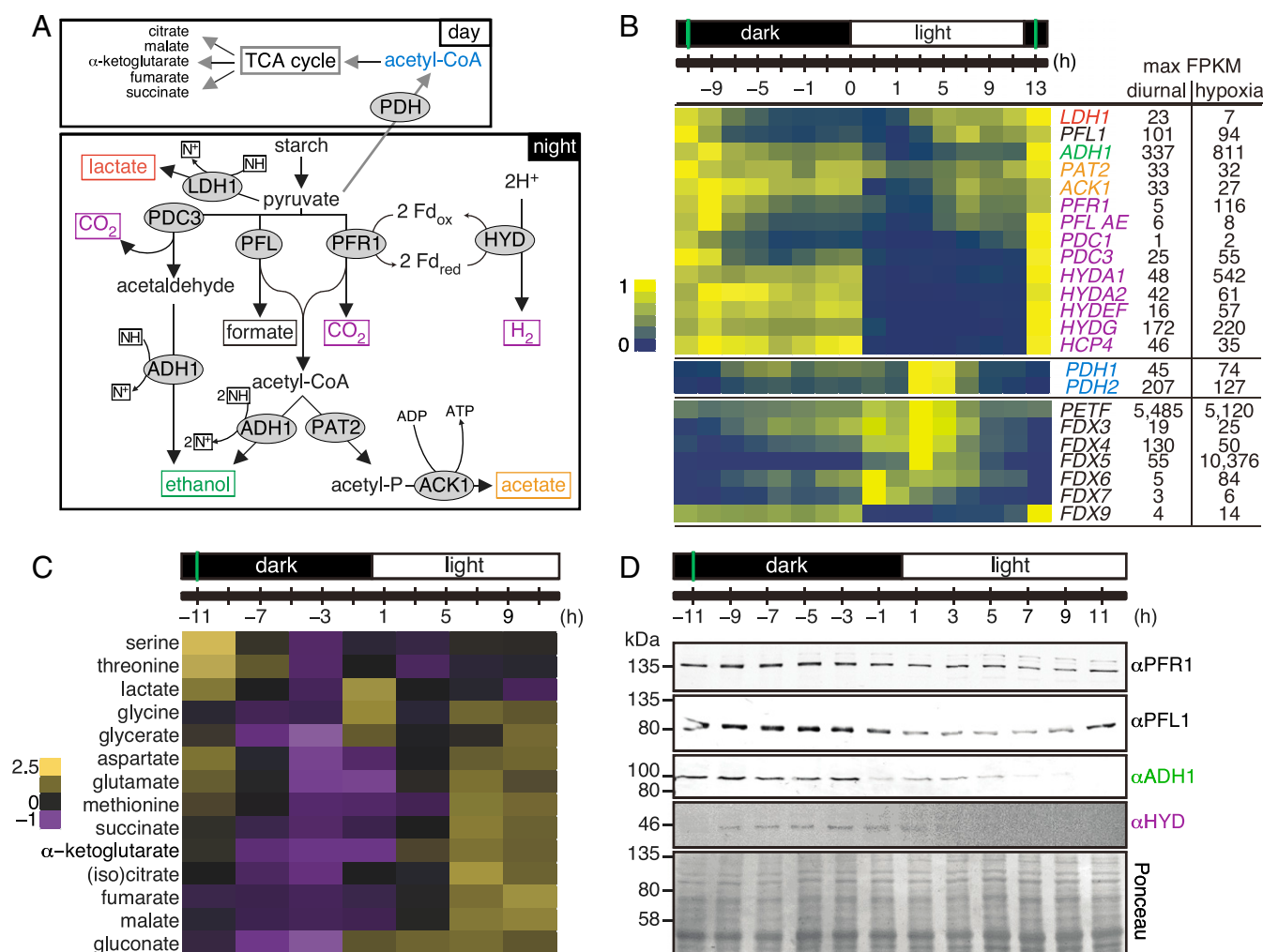


Fig. 6. *Chlamydomonas* cells use anaerobic routes for handling pyruvate. (A) Key pyruvate metabolism pathways according to refs. 36 and 46. Final fermentation products are shown in boxes and enzymes in gray ellipses. Abbreviations: ACK1, acetate kinase 1; ADH1, acetaldehyde/alcohol dehydrogenase; HYDA, [Fe-Fe]-hydrogenase; LDH1, D-lactate dehydrogenase; PAT2, phosphate acetyltransferase 2; PFL1, pyruvate formate lyase; PFR1, pyruvate ferredoxin oxidoreductase. N⁺, NAD⁺; NH, NADH. (B) Normalized expression of fermentation genes listed, shown as a heatmap. Numbers on the right side indicate maximum FPKM values in our samples (diurnal) and in cells grown under dark hypoxia for 6 h (34). (C) Changes in water-soluble metabolites over the diurnal cycle, analyzed by GC-MS. Results are shown as a heatmap of z-score normalized abundance for metabolites that changed significantly over the diurnal cycle. (D) Fermentation enzymes are more abundant in the dark. Total protein samples were separated by denaturing SDS/PAGE, followed by immune-detection with antibodies raised against PFR1, PFL, ADH, and HYDA1+2. Equal protein amounts were loaded and confirmed with Ponceau S stain. All data are shown as average ± SD (n = 3). The immune-detection was performed at least twice on independent samples. The green vertical lines indicate the timing of cell division.

Ferredoxin is the central electron donor in chloroplast metabolism, and the assignment of a particular ferredoxin (of the 13 encoded in the genome) (37, 38) to the hydrogenase is still ambiguous, although *FDX5* is the favored candidate because of its high expression in hypoxia and anaerobiosis. However, *FDX5* shows peak expression during the day (like *PETF*) and is therefore more likely to play a role in photosynthesis (38). Of the seven isoforms predicted to localize to chloroplast (37), only the expression of *FDX9* matches the expression pattern of *HYDA1* and *HYDA2* (Fig. 6B), and may be the more likely candidate.

Acute and Sustained Responses to Light Intensity. Light is the main source of cellular photooxidative damage, and excess light energy may be dissipated as chlorophyll fluorescence or as heat. We were surprised to find signs of light stress in our cultures, although light intensity was moderate and lacked a detectable UV component. Physiological indicators of light stress include a drop in photosystem II efficiency, as measured by *Fv/Fm* (Fig. 7A and *SI Appendix*, Fig. S9B) and a sharp rise in the plastoquinol pool (*SI Appendix*, Fig. S10A and B), concomitant with the induction of the stress-responsive genes encoding light harvesting-like proteins, specifically *LHCSR1*, *LHCSR3s*, and *PSBSs* (Fig. 7D and G and *SI Appendix*, Fig. S9) (39, 40). We only detected this transient induction because of the experimental design, involving CO₂ limitation (air levels) and dense sampling of the culture at dawn, and find it is consistent with the function of PSBS in setting up photoprotection (41). PSBS protein follows *PSBS* expression, with a sharp, transient peak 1 h after lights-on; both messenger and protein are actively degraded within 3 h of light exposure. *LHCSR3* proteins remain constant over the diurnal cycle but are clearly modified upon transfer to light, as expected (*SI Appendix*, Fig. S11) (42).

A build-up of a reduced plastoquinone pool is indicative of a saturation in the plastid ETC between PSII and PSI and overacidification of the plastid lumen, which will promote phosphorylation of LHCII antennae proteins, leading to their dissociation from PSII and association with PSI (43). Measuring chlorophyll fluorescence at 77 °K indeed supports this transition from state I to state II, as evidenced by a higher fluorescence of the PSI-LHC peak at 713 nm (*SI Appendix*, Fig. S10C). The photosynthetic apparatus therefore undergoes a major restructuring in the light to cope with and prevent photooxidative stress, which must occur even at a surface photon flux density of only 200 $\mu\text{mol photons/m}^2/\text{s}$.

Accumulation of *PSBS* mRNA and *PSBS* protein is limited to dawn. Fourteen genes are coexpressed with *PSBS* and include two chlorophyll *a/b*-binding proteins, an E3 ubiquitin ligase and several genes encoding unknown functions (*SI Appendix*, Fig. S12A and B). *LHCSR3* genes show a more prolonged expression window that extends to the middle of the day, where they may function in long-term acclimation to light stress, as evidenced by the partial recovery of *Fv/Fm* parameters (Fig. 7A–C). *LHCSR3* genes share their expression patterns with 21 other genes, among them, a glutaredoxin (*GRX4*) and a fatty acid desaturase (*FAD3*) (*SI Appendix*, Fig. S12C, D, and F). Other genes with strong induction at dawn participate in detoxification of reactive oxygen species, including a glutathione peroxidase (*GPX5*), superoxide dismutase (*FSD1*), and nucleoredoxins (*NRX2* and *NRX3*) (*SI Appendix*, Fig. S12E).

Two Distinct Signaling Pathways Mediate Acclimation to Light Intensity. We reasoned that the expression patterns of *LHCSR3* and *PSBS* genes may reflect a response to (i) light intensity, (ii) the abrupt transition from darkness to light, or (iii) both. We therefore dissected the relative contribution of each signal by adjusting the experimental set-up. A gradual ramping from 0 to 200 $\mu\text{mol photons/m}^2/\text{s}$ over 2 h abrogates *PSBS*'s and *LHCSR3*'s acute induction, but still results in a drop in *Fv/Fm* that is comparable to that seen following an instantaneous transition (Fig. 7D, E, G, and H and *SI Appendix*, Fig. S9). The second peak in *LHCSR3* expression around CT+5 to CT+7 remains and likely represents a response to sustained light intensity (44). Indeed, dropping the fluence rate from 200 to

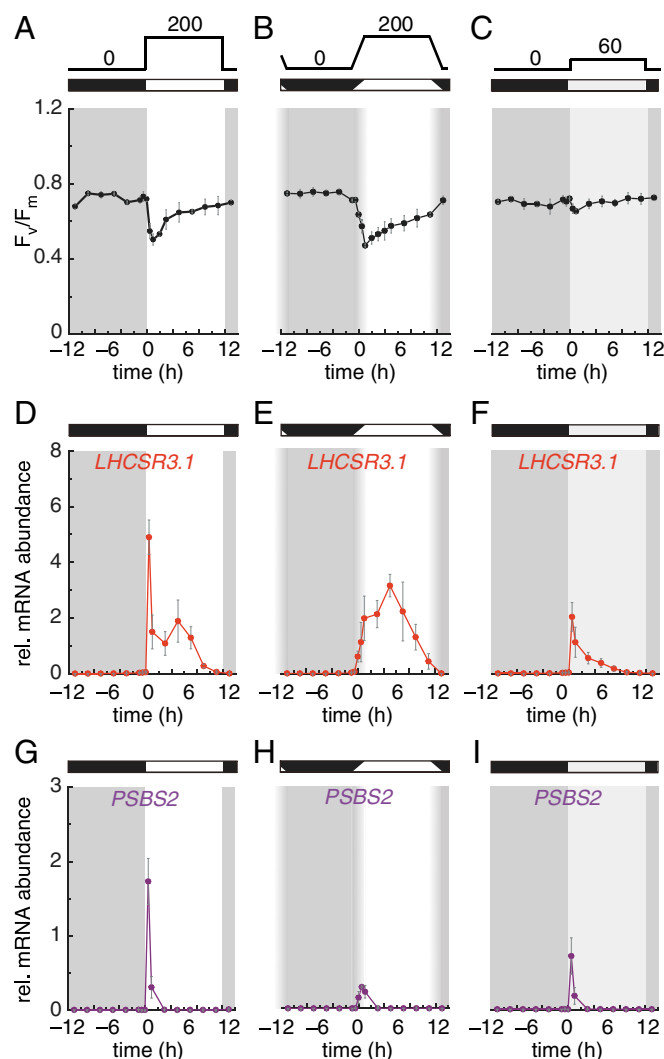


Fig. 7. *Chlamydomonas* cells integrate two light inputs to cope with light stress. (A–C) Photosystem II capacity, as determined by *Fv/Fm* values over the diurnal cycle, in cultures exposed to 200 $\mu\text{mol photons/m}^2/\text{s}$ with abrupt (A) or gradual 2-h transition at dawn (B), or exposed to 60 $\mu\text{mol photons/m}^2/\text{s}$ with abrupt transition at dawn (C). Relative (rel) mRNA abundance for *LHCSR3.1* (D–F) and *PSBS2* (G–I) by quantitative RT-PCR, in samples collected from cells grown under diurnal conditions with 200 $\mu\text{mol photons/m}^2/\text{s}$ with abrupt transition at dawn (D and G), 200 $\mu\text{mol photons/m}^2/\text{s}$ with gradual transition at dawn (E and H), or 60 $\mu\text{mol photons/m}^2/\text{s}$ with abrupt transition at dawn (F and I). All data are shown as average \pm SD ($n = 3$). rel, relative.

60 $\mu\text{mol photons/m}^2/\text{s}$ eliminates the second, later *LHCSR3* peak, confirming our hypothesis (Fig. 7F). Under the lower fluence rate, *Fv/Fm* remains constant and high over the diurnal cycle (Fig. 7C), indicating that cells do not suffer from light stress.

The loss of the later *LHCSR3* peak also uncovers the existence of an acute response at dawn (Fig. 7F), albeit at a lower amplitude than seen when cells are transferred to 200 $\mu\text{mol photons/m}^2/\text{s}$ (Fig. 7D). This acute response is echoed by a peak in *PSBS* expression, although again with a low amplitude (*SI Appendix*, Fig. S9). Cells can therefore measure photosynthetic photon flux over time, and adjust the amplitude of the acute response accordingly.

Discussion

Rhythmic gene expression in the green alga *Chlamydomonas* reflects their biological function and pathways (Fig. 4). Little attention has been given to nighttime physiology, as cells enter

the metabolically quiescent phase (or G_0) of the cell cycle. Our results demonstrate that a greater fraction of pyruvate generated from starch degradation in *Chlamydomonas* enters anaerobic rather than aerobic metabolism. Many fermentation enzymes are oxygen-sensitive, but their transcription is restricted to the night, which is reminiscent of the temporal separation of nitrogen fixation in cyanobacteria (45). For each molecule of pyruvate consumed, *Chlamydomonas* cells therefore produce much less ATP than they could from aerobic metabolism. We estimate that for each ATP produced via fermentation, five are produced via respiration (Fig. 5 and *SI Appendix*), based on the amounts of starch and oxygen consumed at night. Although respiration is still the main source of ATP production at night, we hypothesize that the main contribution of fermentation pathways, aside from ATP itself, is for the reoxidation of NADH. Is this a waste of fuel or is it an indication of unknown metabolic interactions in the natural environment? The products of pyruvate metabolism are typically excreted (46) and may support prokaryotes, which in turn may provide vitamins to the alga (47). Anaerobic pathways can oxidize NADH more rapidly than can oxidative phosphorylation, which may be a consideration at low oxygen tension, a situation that might prevail in the natural environment, and is likely compatible with the low ATP demand during the metabolically more quiescent part of the cell cycle (48).

Like in animals and the related unicellular alga *Volvox*, most *Chlamydomonas* histone transcripts are not poly-adenylated (49, 50) and their expression has therefore been largely unexplored by most RNA-sequencing experiments. Because we used Ribo-depletion, we successfully documented their tight coexpression following DNA replication and cell division, and preceding cilia biogenesis in anticipation of the next dawn and photosynthesis-driven phototaxis and chemotaxis. Of considerable surprise was how greatly histone transcripts are induced: 100- to 1,000-fold, far greater than necessitated by the expected 2-fold increase in protein levels needed to complete mitosis. Their high abundance may, however, drive timely/immediate histone translation. This phenomenon, called RNA superinduction, is becoming recognized as a new regulatory mechanism that allows for competitive translation of particular mRNAs (51). We estimate that almost half of all of the cytosolic ribosomes are busy making more of themselves at dawn by translating mRNAs for ribosomal proteins (*SI Appendix, Table S1*). Regulating protein translation poses a conundrum to a growing cell: how many ribosomes to dedicate to the synthesis of new ribosomal proteins versus all other cellular proteins. Rhythmic expression may have been adopted as an early coping mechanism in unicellular algae: bursts of transcripts at the proper time of day along the diurnal cycle, to produce just enough proteins to make it through another day (52).

Materials and Methods

Strains and Culture Conditions. *Chlamydomonas* strain CC-5390 [CC-4351 (*cw15-325 mt+*) rescued with the pCB412 cosmid carrying the *ARG7* gene] was used for all experiments in this study. Cells were precultivated in 250-mL Erlenmeyer flasks containing 100 mL Tris-acetate-phosphate (TAP) medium with trace element solution, as described previously (53). These precultures were grown in TAP with constant agitation in an Innova incubator (160 rpm, New Brunswick Scientific) at 24 °C in continuous light (90 $\mu\text{mol photons/m}^2/\text{s}$), provided by cool white fluorescent bulbs (4,100 K) and warm white fluorescent bulbs (3,000 K) in a 2:1 ratio until inoculation of photobioreactors.

Photobioreactors Operation and Monitoring. All experiments were performed in presterilized flat-panel photobioreactors (Photobioreactor FMT 150 from Photon System Instruments). Each photobioreactor (working volume 0.4 L) was aerated and mixed with pressurized air at an airflow of 0.2 L/min with mass flow controllers (Smart TMF SLA5850; Brooks). Temperature was set to 28 °C during the day and 18 °C at night, while illumination was provided by a panel of red and blue LEDs (LED Light Source SL 3500; Photon System Instruments). Light fluence was set to 200 $\mu\text{mol photons/m}^2/\text{s}$ unless stated otherwise. For experiments with gradual adjustment of light intensity at dawn and dusk, we used the linear dimming option.

Cells were inoculated from TAP precultures at a starting optical density at 680 nm of 0.05 in high-salt medium (HSM) supplemented with a modified trace element solution (53, 54). Cells were allowed to grow in turbidostat mode for a minimum of 5 d under entraining conditions, until they reached an optical density of 0.4 (corresponding to $2\text{--}3 \times 10^6$ cells/mL). We determined cell number and size with a Beckman Coulter Multisizer 3 with a 50- μm orifice (Beckman Coulter). Samples were concentrated 10-fold in HSM medium before counting. Biological replicates refer to cultures from independent photobioreactors.

RNA Extraction and Library Preparation. A total of 3×10^7 cells were collected at each time-point by centrifugation at $1,424 \times g$ for 5 min at 4 °C. We extracted total RNA with the TRIzol reagent as previously described (55). RNA was DNase-treated with Turbo DNase (Ambion), followed by a cleaning and concentration step with the RNA Clean & Concentrator-5 kit (Zymo Research). RNA quality and concentration were determined on a Nanodrop 2000 (Thermo Fisher Scientific) and an RNA 6000 microfluidic chip on a Bioanalyzer 2100 (Agilent). The University of California, Los Angeles Neuroscience Genomics Core prepared stranded RNA-sequencing libraries using the RiboZero stranded TruSeq RNA Samples prep kit (Illumina). Library quality control was performed by Bioanalyzer with a DNA 1000 microfluidic chip. Libraries were quantified using a Qubit fluorimeter (Thermo Fisher Scientific), pooled and sequenced on a HiSeq 2000 sequencer as single-end 50 bp reads. Transcriptome data were deposited in the NCBI Gene Expression Omnibus (GEO) database under accession GSE112394 (56). Analysis of the resulting data were performed as described previously (16).

Transcriptome Data Analysis.

Multidimensional scaling. Relative expression estimates were imported into the R package *cummeRbund* (56, 57), and subjected to multidimensional scaling with the method *MDSplot* with individual replicates and with mean expression per time point.

k-means clustering and heatmap. Expression estimates for 10,394 nuclear genes (with a Benjamini–Hochberg adjusted *P* value < 0.01 and expression $\geq 1,000$ total counts) were normalized and then subjected to *k*-means clustering with the *kmeans* tool in the R package *amap* with 16 centers. The 16 resulting clusters were arranged manually in order of the peak of expression. Using the same inclusion criteria, 68 chloroplast genes were subjected to *k*-means clustering with three centers. Only three mitochondrial genes met the inclusion criteria. These were normalized and plotted but not subjected to *k*-means clustering. The resulting expression table of clustered genes was normalized by row and plotted as a heatmap with the *heatmap.2* tool in the R package *gplots*.

Phase analysis. The timing (or phase) of peak expression for each gene was determined with the R package *MetaCycle*, using the algorithm JTK_CYCLE (JTK). The three biological replicates were double-plotted to restrict phase-calling to within the diurnal range, set between 20 and 28 h. The JTK method does not allow for uneven sampling intervals or noninteger sampling times; we therefore (i) removed samples collected at CT–0.5 and CT+0.5 and (ii) added columns of missing values at every even-numbered hour, except at time 0, at which time a sample was collected (CT0 being the dark-to light transition). We applied a cut-off of BH.Q of 1×10^{-10} as described previously (8), followed by an expression estimate cut-off of ≥ 1 mean fragments per kilobase of transcript per million mapped reads (FPKM) for at least one sample over the whole time course. Initial estimated phase values were adjusted to the predicted period length of each gene, with the function *adjustPhase* = “predictedPer” within *meta2d* of *MetaCycle*. We followed the same strategy for chloroplast-encoded genes using mean expression across the three replicates, and looked for rhythmic genes with a period of about 24 h (range 20–28). We applied a cut-off of BH.Q of 0.001 (for 24-h rhythms) to account for the single expression estimates used for the analysis. We reanalyzed the Zones et al. dataset (8) with the same settings, keeping all hourly samples and excluding half-hour time points.

Quantitative real-time PCR. Reverse transcription was initiated with an oligo dT₁₈ primer with 2.5 μg total RNA as template and the reverse-transcriptase SuperScript III (Invitrogen), according to the manufacturer’s instructions. The resulting cDNAs were diluted 10-fold before use. Each reaction contained cDNAs corresponding to 100 ng total RNA, 6 pmol each of forward and reverse primers, 0.25 mM dNTPs, 1 \times Ex Taq buffer with Mg^{2+} (TaKaRa), 0.01% (wt/vol) SYBR Green I Nucleic Acid Stain (Cambrex Bio Science Rockland), 0.1% (vol/vol) Tween 20, 100 $\mu\text{g/mL}$ BSA, 5% (vol/vol) DMSO, in a total volume of 20 μL . The following program was used: 95 °C for 5 min, followed by 40 cycles of 95 °C for 15 s, 65 °C for 60 s, with fluorescence measurement after each 65 °C step; a melting curve analysis was performed between 65 and 95 °C to confirm specific amplification. Relative abundances were calculated using LinReg, with EIF1A (Cre02.g103550) as reference transcript. Primer sequences are given in *SI Appendix*.

Proteomics.

Protein detection by LC-MS/MS. We collected 4×10^7 cells by centrifugation at $1,450 \times g$ at 4°C for 4 min. The cell pellet was washed once with 1 mL 10 mM phosphate, pH 7.0, resuspended in 200 μL 10 mM phosphate, pH 7.0 and subjected to slow freeze-thaw cycles to quantitatively extract soluble proteins. Protein concentration of the soluble fraction was determined by BCA assay (Thermo Fisher Scientific). Urea and DTT were added to all samples at a final concentration of 8 M and 10 mM, respectively before incubation at 60°C for 30 min with constant shaking (800 rpm). All samples were then diluted eightfold with 100 mM NH_4HCO_3 and 1 mM CaCl_2 , and digested with sequencing-grade modified porcine trypsin (Promega) provided at a 1:50 (wt/wt) trypsin-to-protein ratio for 3 h at 37°C . Digested samples were desalted using a four-probe positive-pressure Gilson GX-274 ASPEC system (Gilson) with Discovery C18 100-mg/mL solid-phase extraction tubes (Supelco) as follows: columns were pre-conditioned with 3 mL methanol, followed by 2 mL 0.1% trifluoroacetic acid (TFA) in water. Samples were then loaded onto columns, followed by 4 mL 95:5 water:acetonitrile (ACN) 0.1% TFA. Samples were eluted with 1 mL 20:80 water:ACN 0.1% TFA, and concentrated to a final volume of $\sim 100 \mu\text{L}$ in a Speed Vac. After determination of peptide concentration by BCA assay, samples were diluted to 0.25 $\mu\text{g}/\mu\text{L}$ with nanopore water for LC-MS/MS analysis [LC part: LC column of fused silica ($360 \mu\text{m} \times 70 \text{ cm}$) handpacked with Phenomenex Jupiter derivatized silica beads of 3- μm pore size (Phenomenex); HPLC part: HPLC NanoAcquity UPLC system (Waters); MS part: Q Exactive mass spectrometer (Thermo Fisher Scientific)]. Samples were loaded onto LC columns with 0.05% formic acid in water and eluted in 0.05% formic acid in ACN over 100 min. Twelve high-resolution (17.5 K nominal resolution) data-dependent MS/MS scans were recorded for each survey MS scan (35 K nominal resolution) using normalized collision energy of 30, isolation width of 2.0 m/z , and rolling exclusion window lasting 30 s before previously fragmented signals are eligible for reanalysis. Unassigned charge and singly charged precursor ions were ignored.

MS/MS spectra were interrogated against the *Chlamydomonas* proteome via a target-decoy approach; unique peptide sequences with their relative protein abundance as defined by "PeakMaxIntensity" by the in-house MASIS software are given in [Dataset S2](#). The mass spectrometry proteomics data have been deposited in the ProteomeXchange Consortium (www.proteomexchange.org/) under accession no. PXD010794 (58). More details on peak matching and quantification are available in [SI Appendix](#).

Total carbon content analysis. Total nonpurgeable organic carbon content of cells was determined as described previously (59) with minor modifications. We collected 3×10^7 cells by centrifugation at $3,100 \times g$ for 2 min at 4°C , and washed the cell pellet once in 10 mM Phosphate buffer (pH 7.0). Each pellet was then overlaid with 0.9 mL 3 M HCl and digested at 65°C for 16 h with constant agitation. Cell lysates were diluted 111-fold with MilliQ water, for a final hydrochloric acid concentration of 27 mM, and sparged to remove inorganic carbon. Some organic carbon may be purged from the sample by this method, but was not measured in our experimental set-up. We therefore report "nonpurgeable" organic carbon.

Starch measurements. We collected 15-mL culture ($\sim 5 \times 10^7$ cells) by centrifugation at $1,650 \times g$ for 10 min at room temperature, and extracted starch by ethanolic extraction, as previously described (60). We digested starch into glucose with the addition of amyloglucosidase, and then measured glucose levels with the Glucose HK kit (Sigma) according to the manufacturer's instructions.

Immunodetection of *Chlamydomonas* proteins. We collected 15 mL of cultures at a cell density of 2×10^6 cells/mL by centrifugation at $1,650 \times g$ at 4°C . We extracted total proteins by resuspending cell pellets in 300 μL 50 mM Na-phosphate pH 7.0 with cOmplete EDTA-free protease inhibitor (Sigma), 2% (wt/vol) SDS, 10% (wt/vol) sucrose. Protein concentrations were determined with Pierce BCA assay against BSA as standard (Thermo Fisher Scientific). Proteins were separated by SDS/PAGE gels, loading 10 μg protein per lane, and transferred to nitrocellulose membranes by semidry electroblotting. Following blocking in 3% nonfat dried milk in $1 \times$ PBS with 0.1% (wt/vol) Tween 20 for 30 min at room temperature, membranes were incubated with primary and secondary antibodies in the same solution, with intervening washes in $1 \times$ PBS, 0.1% Tween 20. Primary antibodies were used at the following dilutions (with provenance laboratories): hydrogenase 1:2,000 (T. Happe, Ruhr-Universität Bochum, Bochum, Germany); PSBS 1:4,000 (P. Jahns, Heinrich-Heine-Universität in Düsseldorf, Düsseldorf, Germany); PFR 1:3,000 (A. Atteia, Marine Biodiversity, Exploitation and Conservation, Unité Mixte de Recherche, Sète, France); PFL 1:20,000 (A. Atteia); ADH 1:2,500 (A. Atteia); LHCSR3 1:4,000 (M. Hippler, University of Münster, Münster, Germany). A goat anti-rabbit secondary antibody, conjugated to alkaline phosphatase, was used at a dilution of 1:10,000 for detection according to the manufacturer's instructions.

Photosynthetic parameters. Imaging of maximum quantum efficiency of photosystem II was performed using a FluorCam 700 MF system (Photon Systems Instruments) using the Fv/Fm settings as described previously (61). *Chlamydomonas*

strains were dark-adapted for 15 min before each experiment. Fluorescence parameters were calculated as follows: $Fv/Fm = (Fm - Fo)/Fm$, where Fv is the calculated variable fluorescence, Fm is the maximal fluorescence measured immediately after the saturating pulse, and Fo is the initial fluorescence of dark-adapted cells. Fluorescence emission spectra were collected at 77°K , as described previously (62).

Oxygen consumption and evolution measurements. Oxygen evolution rates were measured on a standard Clark-type electrode (Hansatech Oxygraph with a DW-1 chamber) and analyzed with Hansatech OxyLab software v1.15. All experiments were carried out on 2 mL of cells removed from the culture (density of $\sim 2\text{--}3 \times 10^6$ cells/mL) in the presence of 10 mM KHCO_3 and under constant stirring. Respiration rates were measured as oxygen consumption over a period of at least 5 min in the dark (or until the observed rate stabilizes). We measured oxygen evolution on the same samples over 5 min by turning the built-in LED to provide 200 $\mu\text{mol photons}/\text{m}^2/\text{s}$ (as in the photobioreactors) and 500 $\mu\text{mol photons}/\text{m}^2/\text{s}$ (saturating light) after a 5-min dark acclimation period. The rate of photosynthetic oxygen evolution was calculated as the difference between oxygen evolution in the light and oxygen consumption in the dark for each sample.

For inhibitor studies, we prepared fresh stock solutions of all chemicals in 100% ethanol [100 mM *n*-propyl gallate, Sigma P3130; 400 mM salicylhydroxamic acid (SHAM), Sigma-Aldrich S7504; 10 mM carbonyl cyanide 4-(trifluoromethoxy)phenylhydrazone (FCCP), Sigma-Aldrich C2920; or MilliQ water (200 mM potassium cyanide, Fisher P223-100)]. Final concentrations for inhibitors during the experiments were 1 mM *n*-Propyl gallate, 5 mM SHAM, 4 mM potassium cyanide and 5 μM FCCP (for respiratory capacity), and respiration rate was measured as oxygen consumption over 5 min in the dark, and again after an acclimation period in the presence of the inhibitors. Control samples received ethanol alone at a concentration of 1% to account for potential solvent effects.

Metabolites analysis by GC-MS. Metabolites extraction was performed as previously described (63). Briefly, cells were collected from 10 mL of a culture that had reached a cell density of $2\text{--}4 \times 10^6$ cells/mL by vacuum filtration onto a PVDF filter (GVWP02500; Millipore). Filters were frozen immediately in liquid nitrogen. Extraction of metabolites was carried out by soaking filters in 600 μL of cold 70% methanol 30% chloroform (vol/vol) with vortexing at 4°C for 70 min. Filters were removed, and samples collected by a short centrifugation before the addition of 300 μL cold LC-grade water, followed by two freeze-thaw cycles. After centrifugation, the aqueous polar phase was collected and dried in a Speed Vac. Derivatization and analysis of metabolites was performed as described (64) by gas chromatography-mass spectrometry (GC-MS) on a 7200 GC-QTOF instrument (Agilent). Data analysis was conducted with the help of the Mass Hunter Software (Agilent). For relative normalization, all metabolite peak areas were normalized to cell number.

Plastoquinone measurements. We collected a total of $4\text{--}8 \times 10^7$ cells/mL *Chlamydomonas* cells by centrifugation at $1,424 \times g$ for 5 min at 4°C . Cell pellets were resuspended in 500 μL 95% (vol/vol) ethanol spiked with 50 μL of 121 μM ubiquinone-10 (final amount of 6.05 nmols) as internal standard, and homogenized in a 5-mL Pyrex tissue grinder. The grinder was rinsed with 500 μL 95% (vol/vol) ethanol, and combined with the initial homogenate. We separated cell debris by centrifugation at $18,000 \times g$ for 5 min at 4°C and immediately analyzed extracts by HPLC as described using 100 μL of each extract (65). We tested whether the centrifugation step may affect plastoquinol/plastoquinone ratios in the samples by measuring plastoquinol and plastoquinone in matched sets with either centrifugation or direct quenching by the addition of ethanol ($\sim 70\%$ final concentration) to cell cultures. Plastoquinol/plastoquinone ratios were not significantly different between sets, indicating that the redox state of plastoquinone is not affected during sample collection under our conditions.

Data availability. Transcriptome data were deposited at the National Center for Biotechnology Information's Gene Expression Omnibus under accession GSE112394 (56).

The mass spectrometry proteomics data have been deposited to the ProteomeXchange Consortium via the PRIDE (66) partner repository with the dataset identifier PXD010794 and 10.6019/PXD010794 (58).

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